

Adaptation in plant genomes: bigger is different

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Since their origin 160 million years ago, flowering plants have rapidly diversified into more than 300,000 species, adapting to a striking array of habitats and conditions. In this time, flowering plants have dominated some of the most diverse and extreme environments, harnessed a number of specialized biotic interactions in order to ensure successful pollen and seed dispersal, and adapted to meet the demands of agriculture. Given their diversity and importance, a considerable body of research has been devoted to understanding plant adaptation (Tiffin and Ross-Ibarra, 2014), but the relative importance of the various factors that may impact the process of adaptation are still not well understood. For instance, transitions in polyploidy and mating system have long been considered plausible drivers of flowering plant adaptation (Soltis et al., 2009; Goldberg and Igić, 2012), but both are also associated with evolutionary dead ends (Mayrose et al., 2011; Igić and Busch, 2013). As another example, while the effective size of plant populations is expected to correlate with estimates of the efficiency of natural selection, empirical support for this prediction is mixed (Strasburg et al., 2010; Gossmann et al., 2010). A number of aspects of adaptation have received a degree of theoretical (Hermisson and Pennings, 2017; Ralph and Coop, 2010) or empirical (Strasburg et al., 2012; Anderson et al., 2013; Ågren et al., 2013; Yeaman et al., 2016) support, but it is clear we are far from understanding all of the factors underlying the process of adaptation in flowering plants.

Here we propose that genome size may play a previously under-appreciated role in determining how plants adapt. Rather than focus on the mechanisms of genome size variation (Lynch et al., 2016; Lynch and Conery, 2003) or the adaptive significance of genome size itself (Grime and Mowforth, 1982; Bilinski et al., 2017), our **functional space hypothesis** predicts that interspecific differences in genome size may affect the process of adaptation by changing the number and location of potentially functional mutations. Below, we outline an argument for why existing differences in genome size — spanning more than three orders of magnitude across flowering plants (Gaut and Ross-Ibarra, 2008) — may lead to differences in “functional space”, and what this implies about adaptation in plant genomes.

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36 Larger genomes have more functional space

37 Mutation rates are typically reported at the level of individual nucleotides, but unless these
38 rates change in larger genomes, plants with more nucleotides will inevitably be subject to
39 more mutations *per genome* in each generation. And because variation in gene number across
40 plants is not substantial (Bennetzen et al., 2005), most of the additional mutational input
41 in larger genomes occurs outside of coding sequence.

42
43 But not only nucleotides mutate. By definition, large genomes are a consequence of inser-
44 tion of additional base pairs. In plants, diploid genome size expansion is often the result of
45 amplification of transposable elements (TEs). In addition to the nucleotides contributed by
46 their insertion, the presence of TEs can generate variation in gene expression and regulation
47 and even generate genome rearrangements (Chuong et al., 2016; Lisch, 2013). Beyond TEs,
48 structural variation such as inversions are likely to be abundant in large genomes due to
49 their association with unstable intergenic DNA. Inversions have often been found to play
50 an oversized role in plant adaptation, from perenniality in monkeyflower (Lowry and Willis,
51 2010) to flowering time in maize (Navarro et al., 2017) and *Boechera* (Lee et al., 2017) to
52 fecundity under drought in *Arabidopsis* (Fransz et al., 2016). Other structural variation is
53 likely more common in larger genomes as well. Gene movement is frequent in large genomes
54 such as wheat and barley (Wicker et al., 2011), frequently due to the action of transposable
55 elements (Morgante et al., 2005). Variation in gene copy number is also common (Żmieńko
56 et al., 2014), and has been identified as the source of a number of different adaptations
57 (Pham et al., 2017; Prunier et al., 2017).

58
59 Together, nucleotide changes and structural changes combine to contribute an increased
60 influx of new variation in large plant genomes. Because some portion of these mutations
61 will have phenotypic consequences, this new variation provides greater mutational “space”
62 on which selection can act. While clearly most mutational changes are unlikely to impact
63 function, especially in intergenic regions, if even a small fraction of this mutational input in
64 larger genomes impacts phenotype, selection can act to maintain or remove these mutations
65 in populations. Indeed, the quantity of expressed intergenic sequence appears to scale with
66 genome size (Lloyd et al., 2017), and machine learning approaches predict a substantial mi-
67 nority of these sequences are likely functional (Lloyd et al., 2017). And though evidence of
68 functionality is difficult to come by for most species, substantially more loci associated with
69 phenotypic variation are found in intergenic regions far from genes (Figure 1) in the large
70 genome of maize than the small *Arabidopsis* genome.

71
72 We envision that new mutations in intergenic regions are most likely to impact function if
73 they affect regulatory sequence. Regulatory sequences across the genome can be identified
74 by their signature of open chromatin via nuclease (MNase-seq or DNase-seq) or transposase
75 (ATAC-seq) accessibility (e.g. Oka et al., 2017). Evidence that open chromatin may be a
76 useful proxy for functional sequence can be found in maize, where open chromatin makes
77 up less than 1% of the genome but nucleotide variation in such regions explains more than
78 40% of phenotypic variation across traits (Rodgers-Melnick et al., 2016). For the additional
79 mutational input in larger genomes to be functional, we expect to find more of the intergenic

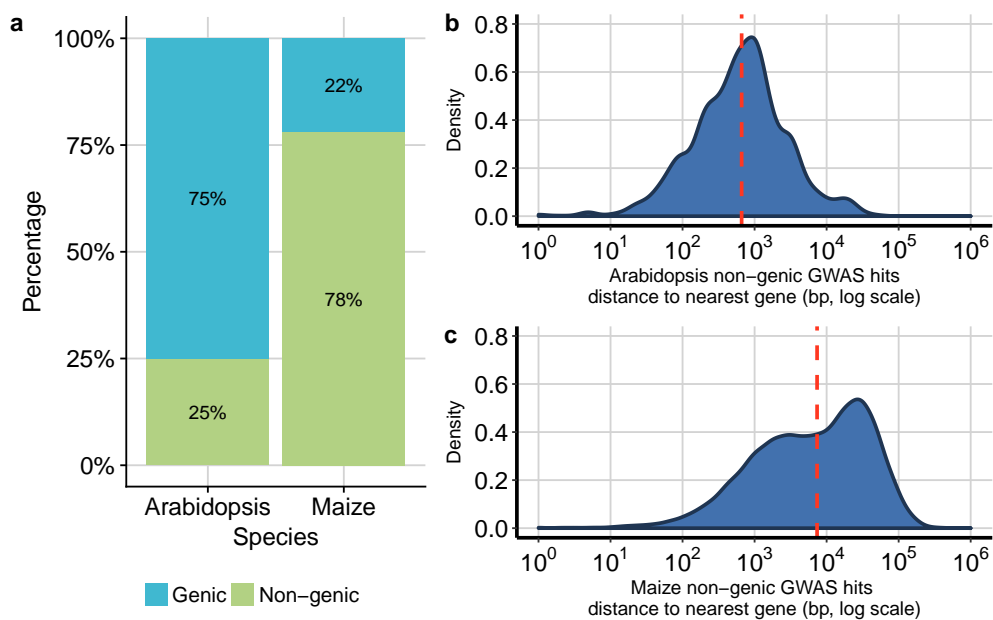


Figure 1: **Genomic distribution of loci identified in genome-wide association studies in *Arabidopsis* and maize.** *Arabidopsis* GWAS hits are top associations from the AraGWAS catalog (<https://aragwas.1001genomes.org>), and maize GWAS hits are the curated results of the nested association mapping population from maize diversity project (<https://www.panzea.org>) (Wallace et al., 2014). (a) Proportion of significant GWAS hits in genic (exon and intron) and non-genic regions. (b-c) Density plots of non-genic GWAS hits in (b) *Arabidopsis* and (c) maize. Dotted red lines indicate medians.

80 sequence in large genomes as open chromatin and "functional space". In line with these
81 expectations, Maher et al. (2017) show that total accessible chromatin space increases as
82 genome size increases across four species ranging almost an order of magnitude in genome
83 size ($p=0.11$, adjusted $R^2 = 0.11$; Figure 2a), and that the largest increase in chromatin
84 between species is in intergenic regions. Although chromatin accessibility data prepared in
85 different labs and from different tissues makes direct comparison difficult, aggregated data
86 from a broad range of species is also suggestive of increases in non-exonic open chromatin as
87 genome size increases (Figure 2b).

88

89 In total, larger genomes have more mutational input, most of which occurs in intergenic
90 regions. Some proportion of this mutational input leads to additional open chromatin,
91 increasing the "functional space" of large genomes.

92 Large genomes adapt primarily through mutations in 93 regulatory regions

94 One common approach to study adaptation polarizes synonymous and nonsynonymous sub-
95 stitutions using sequence comparison to an outgroup. Because nonsynonymous mutations
96 are more likely to be functional, an observed excess of nonsynonymous substitutions com-
97 pared to expectations from synonymous sites likely reflects the effects of positive selection.

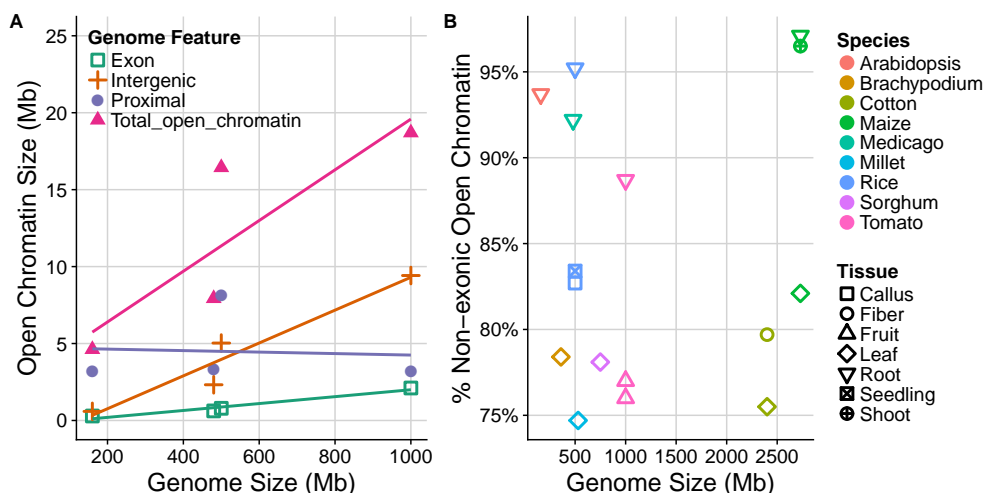


Figure 2: **Open chromatin and genome size.** (a) The total amount of open chromatin scales with genome size in four species *Arabidopsis thaliana*, *Medicago truncatula*, *Solanum lycopersicum* and *Oryza sativa* (Maher et al., 2017). Intergenic regions are defined >2kb from the transcription start site and >1kb from the transcription termination site. (b) Percentage of open chromatin outside of exons scales with genome size across many species. References and details of analyses for this figure can be found at https://github.com/RILAB/AJB_MutationalTargetSize_GenomeSize

98 The proportion of adaptive nonsynonymous substitutions (α) is used to infer the strength
 99 or efficiency of natural selection (Eyre-Walker, 2006). But as most of the additional func-
 100 tional mutations in large genomes occur in regulatory regions outside of genes (Figure 2), as
 101 genomes increase in size, a larger proportion of adaptive variation should occur outside genes
 102 and enrichment for nonsynonymous substitutions should decrease. Limited evidence for this
 103 effect can be found in two studies of environmental adaptation, where putatively selected
 104 loci are enriched for nonsynonymous mutations in the small *Arabidopsis* genome (Hancock
 105 et al., 2011) but for noncoding sequence near genes in the larger teosinte genome (Pyhäjärvi
 106 et al., 2013). We thus predict that, all else being equal, plants with larger genomes should
 107 exhibit lower α values than species with smaller genomes. After correcting for phylogeny,
 108 we indeed observe such a negative slope ($p=0.12$; Figure 3), and a negative correlation is
 109 also evident within the only two well-sampled genera available (*Helianthus* and *Pinus*). The
 110 effect we predict may be relatively weak overall, however, as a number of other factors such
 111 as effective population size and population structure (Gossmann et al., 2010) have also been
 112 suggested to impact α . Published data of α for species with small genome size are biased to-
 113 wards species with a small effective population size — and thus less efficient natural selection
 114 — which might influence the correlation between α and genome size in our meta-analysis
 115 (Figure 3).

116 Sweeps are softer in larger genomes

117 At the onset of a new selective pressure, adaptation can either proceed using genetic varia-
 118 tion currently segregating in the population or by acting upon *de novo* beneficial mutations

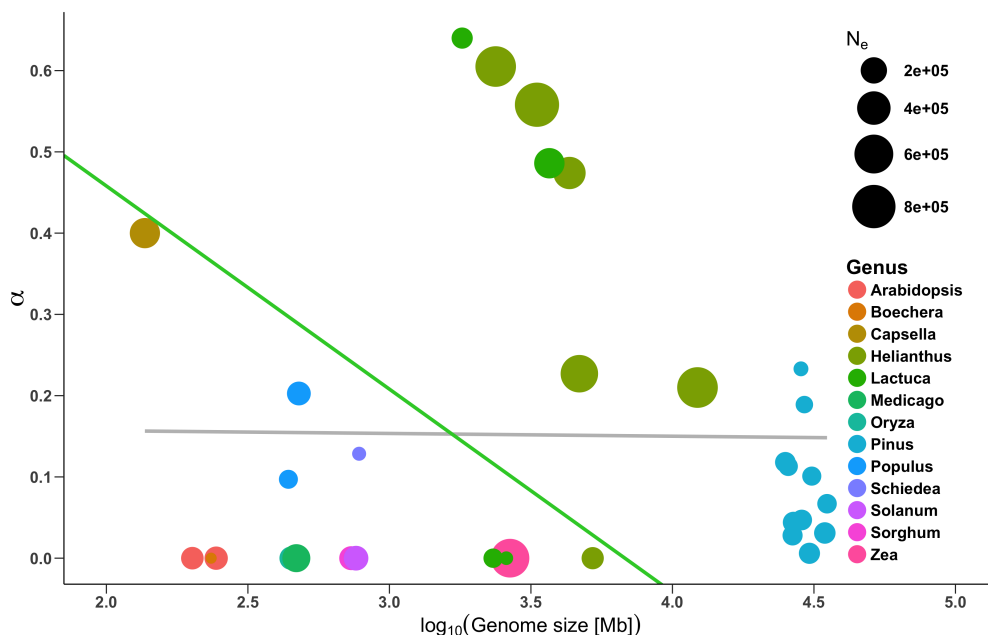


Figure 3: **Pattern of adaptive substitutions and genome size.** The proportion of adaptive substitutions (α) across plants with different genome sizes. Colors represent genera and point sizes show estimates of the effective population size (N_e). Lines represent regression analyses performed with (green) and without (black) phylogenetic correction. References and details of analyses for this figure can be found at https://github.com/RILAB/AJB_MutationalTargetSize_GenomeSize

119 as they enter the population. If mutations are limited, adaptation occurs predominantly by
 120 new beneficial mutations, which rapidly reach high frequency and leave a footprint of reduced
 121 local genetic diversity known as a “hard sweep” (Smith and Haigh, 1974). A number of other
 122 modes of adaptation are possible, however, and often collectively referred to as “soft sweeps”.
 123 If substantial functional variation currently exists in the population, adaptation can make
 124 use of existing variants and affect surrounding genetic diversity to a lesser degree (Hermisson
 125 and Pennings, 2017). Similarly, if the influx of new beneficial mutations across multiple
 126 loci is sufficiently high, adaptation can occur via partial sweeps that increase the frequency
 127 — but not fix — multiple independent beneficial mutations (Hermisson and Pennings, 2017).

128
 129 A key component in determining which of these processes take place as adaptation progresses
 130 is the mutational target size — what we term here more generally as “functional space”. More
 131 formally, the likelihood of sweeps being hard or soft depends on the population mutation
 132 rate θ_b , which is the product of twice the effective population size (N_e) and the genome-wide
 133 beneficial mutation rate (U_b) (Hermisson and Pennings, 2017). While there is little evidence
 134 that the effective size of plant populations scales with genome size (Whitney et al., 2010),
 135 as we argue above the genome-wide rate of beneficial mutations should increase as genomes
 136 increase in size. Theory suggests that soft sweeps should predominate whenever $\theta_b > 1$
 137 (Hermisson and Pennings, 2017), leading to the prediction that adaptation from extant
 138 variation or multiple independent mutations should be more common in larger genomes.
 139 Although empirical examinations of the amount of soft and hard sweeps across populations

140 is still scarce, patterns of diversity around nonsynonymous substitutions in the small *Capsella*
141 *grandiflora* genome suggests a predominance of hard sweeps (Williamson et al., 2014), while
142 similar analyses in the larger maize genome find little evidence for hard sweeps around
143 nonsynonymous substitutions (Beissinger et al., 2016).

144 **Functional space and adaptation in plant genomes**

145 Here we have proposed the **functional space hypothesis**, positing that mutational target
146 size scales with genome size, impacting the number, source, and genomic location of beneficial
147 mutations that contribute to adaptation. Though motivated by preliminary evidence, mostly
148 from *Arabidopsis* and maize, more data are needed before any rigorous assessment of the
149 hypothesis can be made. If correct, the functional space hypothesis suggests that we should
150 expect plants with large genomes to exhibit more functional mutations outside of genes,
151 more regulatory variation, and likely less signal of strong selective sweeps reducing diversity.
152 These differences have implications for how we study the evolution and development of
153 plant genomes, from where we should look for signals of adaptation to what patterns we
154 expect adaptation to leave in genetic diversity or gene expression data. While flowering
155 plant genomes vary across more than three orders of magnitude in size, most studies of both
156 functional and evolutionary genomics have focused on species at the extreme small edge of
157 this scale. Our hypothesis predicts that methods and results from these small genomes may
158 not replicate well as we begin to explore large plant genomes. Finally, while we have focused
159 here on evidence from plant genomes, we see no *a priori* reason why similar arguments might
160 not hold in other taxa as well.

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